

Question 1

(a)

When $n=1$:

$$P(\theta | x_1=1) = P(\theta) \frac{P(x_1=1 | \theta)}{P(x_1=1)}$$

$$= P(\theta) \frac{P(x_1=1 | \theta)}{\int_0^1 P(x_1=1 | \theta) P(\theta) d\theta}$$

$$= P(\theta) \frac{\theta}{\int_0^1 \theta P(\theta) d\theta}$$

$$\dots \dots \dots P(x=1 | \theta=0) = 0$$

$$= P(\theta) \frac{\theta}{E[\theta]}$$

$$\dots \dots \dots E[f(x)] = \int f(x) P(x) dx.$$

when $n=n$:

Assumes that $P(\theta | x_1=n=1^n) = P(\theta) \frac{\theta^n}{E[\theta^n]}$

To prove:

$$P(\theta | x_1=n+1=1^{n+1}) = P(\theta) \frac{\theta^{n+1}}{E[\theta^{n+1}]}$$

$$P(\theta | x_{1:n} = \{u\})$$

$$= P(\theta | x_{1:n} = \{u\}) \frac{P(x_{n+1} = 1 | \theta, x_{1:n} = \{u\})}{P(x_{n+1} = 1 | x_{1:n} = \{u\})}$$

$$= P(\theta | x_{1:n} = \{u\}) \frac{P(n+1 = 1 | \theta, x_{1:n} = \{u\})}{\int_0^1 P(x_{n+1} = 1 | \theta, x_{1:n} = \{u\}) d\theta}$$

$$= P(\theta | x_{1:n} = \{u\}) \frac{P(n+1 = 1 | \theta, x_{1:n} = \{u\})}{\int_0^1 P(x_{n+1} = 1 | \theta, x_{1:n} = \{u\}) P(\theta | x_{1:n} = \{u\}) d\theta}$$

$$= P(\theta | x_{1:n} = \{u\}) \frac{\theta}{\int_0^1 \theta \cdot P(\theta | x_{1:n} = \{u\}) d\theta}$$

$$\dots \dots P(x_{n+1} = 1 | \theta, x_{1:n} = \{u\}) = 1$$

$$= \frac{\theta^n P(\theta)}{E[\theta^n]} \frac{\theta}{\int_0^1 \theta \frac{\theta^n P(\theta)}{E[\theta^n]} d\theta}$$

$$= P(\theta) \frac{\theta^{n+1}}{\int_0^1 \theta^{n+1} P(\theta) d\theta}$$

$$= P(\theta) \frac{\theta^{n+1}}{E[\theta^{n+1}]}$$

$$\text{Therefore, } P(\theta | x_{1:n} = \{u\}) = P(\theta) \frac{\theta^n}{E[\theta^n]}$$

(b)

When $n=0$:

$$\begin{aligned} & P(\theta | x_1=1) \\ & = P(\theta) \frac{P(x_1=0|\theta)}{P(x_1=0)} \\ & = P(\theta) \frac{P(x_1=0|\theta)}{\int_0^1 P(x_1=0|\theta) p(\theta) d\theta} \\ & = P(\theta) \frac{1-\theta}{\int_0^1 (1-\theta) p(\theta) d\theta} \quad \dots \dots \quad P(x_1=1|\theta=0)=\theta \\ & = P(\theta) \frac{1-\theta}{E[1-\theta]} \quad \dots \dots E[f(x)] = \int f(x) p(x) dx. \end{aligned}$$

when $n=n$:

Assumes that $P(\theta | x_1=n=0^n) = P(\theta) \frac{(1-\theta)^n}{E[(1-\theta)^n]}$

To prove:

$$P(\theta | x_{1:n+1}=0^{n+1}) = P(\theta) \frac{(1-\theta)^{n+1}}{E[(1-\theta)^{n+1}]}$$

$$P(\theta | x_{1:n+1}=0^{n+1})$$

$$= P(\theta | x_{1:n} = 0^n) \frac{P(x_{n+1} = 0 | \theta, x_{1:n} = 0^n)}{P(x_{n+1} = 0 | x_{1:n} = 0^n)}$$

$$= P(\theta | x_{1:n} = 0^n) \frac{P(x_{n+1} = 0 | \theta, x_{1:n} = 0^n)}{\int_0^1 P(x_{n+1} = 0 | \theta, x_{1:n} = 0^n) d\theta}$$

$$= P(\theta | x_{1:n} = 0^n) \frac{P(x_{n+1} = 0 | \theta, x_{1:n} = 0^n)}{\int_0^1 P(x_{n+1} = 0 | \theta, x_{1:n} = 0^n) P(\theta | x_{1:n} = 0^n) d\theta}$$

$$= P(\theta | x_{1:n} = 0^n) \frac{1 - \theta}{\int_0^1 (1 - \theta) P(\theta | x_{1:n} = 0^n) d\theta}$$

$$= \frac{(1 - \theta)^n P(\theta)}{E[(1 - \theta)^n]} \frac{1 - \theta}{\int_0^1 (1 - \theta)^n P(\theta) d\theta} = \dots \cdot P(x_{n+1} = 1 | \theta, x_{1:n} = 0^n) = 1$$

$$= P(\theta) \frac{(1 - \theta)^{n+1}}{E[(1 - \theta)^{n+1}]}$$

$$\text{Therefore, } P(\theta | x_{1:n} = 0^n) = P(\theta) \frac{(1 - \theta)^n}{E[(1 - \theta)^n]}$$

(c).

$$P(\theta | x_{1:n} = \cdot^n) = P(\theta) \frac{\theta^n}{\int_0^1 \theta^n p(\theta) d\theta}$$

For $\int_0^1 \theta^n p(\theta) d\theta$:

$$\int_0^1 \theta^n d\theta = \frac{\theta^{n+1}}{n+1} \Big|_0^1 = \frac{1}{n+1}$$

Therefore, $P(\theta | x_{1:n} = \cdot^n) = (n+1) \theta^n$

(d)

$$P(\theta | x_{1:n} = \cdot^n) = (n+1) \theta^n \quad \dots \text{From Q1.C}$$

$$\begin{aligned} M_n &= \int_0^1 \theta \cdot P(\theta | x_{1:n} = \cdot^n) d\theta \\ &= \int_0^1 (n+1) \theta^{n+1} d\theta \\ &= (n+1) \cdot \frac{\theta^{n+2}}{n+2} \Big|_0^1 \\ &= \frac{n+1}{n+2} \end{aligned}$$

When $n \rightarrow \infty$:

$$\lim_{n \rightarrow \infty} \frac{n+1}{n+2} = \lim_{n \rightarrow \infty} \frac{1 + \frac{1}{n}}{1 + \frac{2}{n}} = \frac{1 + \lim_{n \rightarrow \infty} (\frac{1}{n})}{1 + \lim_{n \rightarrow \infty} (\frac{2}{n})} = 1$$

Therefore, when $n \rightarrow \infty$, $M_n \rightarrow 1$.

(e).

$$P(\theta | x_{1:n} = \cdot^n) = (n+1) \theta^n \quad \dots \text{From Q1.C}$$

$$\begin{aligned} S_n^2 &= \int_0^1 (\theta - M_n)^2 \cdot P(\theta | x_{1:n} = \cdot^n) d\theta \\ &= \int_0^1 (n+1) \left(\theta - \frac{n+1}{n+2} \right)^2 \theta^n d\theta \\ &= \int_0^1 (n+1) \left(\theta^2 + \left(\frac{n+1}{n+2} \right)^2 - \frac{2\theta(n+1)}{n+2} \right) \theta^n d\theta \\ &= \int_0^1 \left((n+1)\theta^2 + \frac{(n+1)^3}{(n+2)^2} - \frac{2\theta(n+1)^2}{n+2} \right) \theta^n d\theta \\ &= \int_0^1 \left((n+1)\theta^{n+2} + \frac{(n+1)^3}{(n+2)^2} \theta^{n-2} - \frac{2(n+1)^2}{n+2} \theta^{n+1} \right) d\theta \\ &= \left[\frac{n+1}{n+3} \theta^{n+3} + \frac{(n+1)^2}{(n+2)^2} \theta^{n+1} - \frac{2(n+1)^2}{(n+2)^2} \theta^{n+2} \right] \Big|_0^1 \\ &= \frac{n+1}{n+3} + \frac{(n+1)^2}{(n+2)^2} - \frac{2(n+1)^2}{(n+2)^2} \end{aligned}$$

When $n \rightarrow \infty$:

$$G_n^2 = \frac{(n+1)(n+2)^2}{(n+3)(n+2)^2} - \frac{(n+1)^2(n+3)}{(n+2)^2(n+3)}$$

$$\begin{aligned} &= \frac{(n+1)(n+2)^2 - (n+1)^2(n+3)}{(n+3)(n+2)^2} \\ &= \frac{\frac{n+1}{n+3} \cdot \frac{(n+2)^2}{n+3} - \frac{(n+1)^2}{n+3}}{(n+2)^2} \end{aligned}$$

$$\therefore (n+2)^2 > \frac{n+1}{n+3} \cdot \frac{(n+2)^2}{n+3}, \quad (n+2)^2 > \frac{(n+1)^2}{n+3}$$

\therefore when $n \rightarrow \infty$, $G_n^2 \rightarrow 0$

(f)

$$P(\Theta | X_{1:n} = \theta^n) = (n+1) \theta^n \quad \dots \dots \text{From Q.I.C}$$

For $\Theta \in (0, 1]$:

$$P'(\Theta | X_{1:n} = \theta^n) = n(n+1) \theta^{n-1} > 0$$

Then the maximum Θ MAP_n is 1.

For any $\theta \in (0, 1)$ has the same probability.

Therefore. $\Theta_{MAP} = 1$ for any $\theta \in (0, 1)$.

(g)

Because for any $\theta \in (0, 1)$ has the same probability. The best guess of θ , M_n and Θ_{MAP} is 1.

(h)

$$P(\theta | x_{1:n} \sim \cdot^n) = (n+1) \theta^n \quad \dots \text{From Q1.c}$$

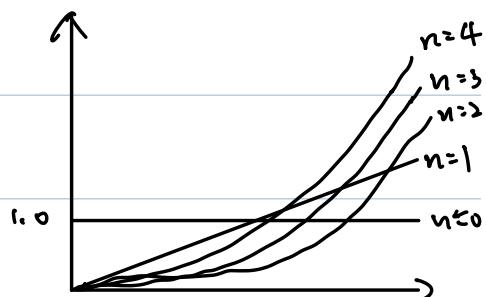
$$n=0: P(\theta | x_{1:0} \sim \cdot^0) = 1$$

$$n=1: P(\theta | x_{1:1} \sim \cdot^1) = 2\theta$$

$$n=2: P(\theta | x_{1:2} \sim \cdot^2) = 3\theta^2$$

$$n=3: P(\theta | x_{1:3} \sim \cdot^3) = 4\theta^3$$

$$n=4: P(\theta | x_{1:4} \sim \cdot^4) = 5\theta^4$$



Therefore, $P(\theta)$ approaching 1 with increase of n .

Question 2

(a)

For $\alpha = \beta = 1$:

when the win is 0, consumer's result is 0.

Therefore, the agent updates its posterior approaching 0.

when the win is 1, consumer's result is 1.

Therefore, the agent updates its posterior approaching 1.

For $\alpha = \beta = 1/2$:

Whatever the coin is 0 or 1, consumer's result has

50% to get correct answer, which means

the posterior will not change.

Therefore, the agent will not update its posterior.

For $\alpha = \beta = 0$:

when the van is 0, camera's result is 1.

Therefore, the agent updates its posterior approaching 0.

when the van is 1, camera's result is 0.

Therefore, the agent updates its posterior approaching 1.

(b)

$$\begin{aligned} & P(\hat{X}=x|\Theta) \\ &= \sum_{n=0}^1 P(\hat{X}=x, X=n|\Theta) \\ &= \sum_{n=0}^1 P(\hat{X}=x|X=n, \Theta) \cdot P(X=n|\Theta) \\ &= P(\hat{X}=x|X=0, \Theta)P(X=0|\Theta) + P(\hat{X}=x|X=1, \Theta)P(X=1|\Theta) \end{aligned}$$

we have:

$$P(\hat{X}=0 | X=0) = \alpha \dots \textcircled{1}$$

$$P(\hat{X}=0 | X=1) = 1-\beta \dots \textcircled{2}$$

$$P(\hat{X}=1 | X=0) = \beta \dots \textcircled{3}$$

$$P(\hat{X}=1 | X=1) = 1-\alpha \dots \textcircled{4}$$

Base on ② and ③:

Because $x=0$ is the result of θ for $P(\hat{x}=x|x=0, \theta)$,
and $x=1$ is the result of θ for $P(\hat{x}=x|x=1, \theta)$.

$$P(\hat{x}=x|\theta)$$

$$\begin{aligned} &= P(\hat{x}=x|x=0, \theta)(1-\theta) + P(\hat{x}=x|x=1, \theta)\theta \\ &= P(\hat{x}=x|x=0)(1-\theta) + P(\hat{x}=x|x=1)\theta \end{aligned}$$

when $x=0$:

$$P(\hat{x}=0|\theta)$$

$$\begin{aligned} &= P(\hat{x}=0|x=0)(1-\theta) + P(\hat{x}=0|x=1)\theta \\ &= 2(1-\theta) + (-\beta)\theta \quad \cdots \text{--- } ① \text{ and } ② \end{aligned}$$

when $x=1$:

$$P(\hat{x}=1|\theta)$$

$$\begin{aligned} &= P(\hat{x}=1|x=0)(1-\theta) + P(\hat{x}=1|x=1)\theta \\ &= (1-\beta)(1-\theta) + \beta\theta \quad \cdots \text{--- } ③ \text{ and } ④ \end{aligned}$$

(c)

Base on Bayes theorem:

$$P(\theta | \hat{x}=1) = P(\theta) \cdot \frac{P(\hat{x}=1 | \theta)}{P(\hat{x}=1)}$$

$$= P(\theta) \cdot \frac{P(\hat{x}=1 | \theta)}{\int_0^1 P(\hat{x}=1 | \theta) P(\theta) d\theta}$$

We have:

$$P(\hat{x}=1 | \theta) = (1-\theta)(1-\theta) + \beta\theta \quad \dots \text{From Q2.b}$$

$$P(\theta | \hat{x}=1) = P(\theta) \cdot \frac{(1-\theta)(1-\theta) + \beta\theta}{\int_0^1 ((1-\theta)(1-\theta) + \beta\theta) P(\theta) d\theta}$$

When $\alpha=\beta=1$

$$\begin{aligned} P(\theta | \hat{x}=1) &= P(\theta) \frac{\theta}{\int_0^1 \theta P(\theta) d\theta} \\ &= P(\theta) \frac{\theta}{E[\theta]} \quad \dots \text{--- } E[f(x)] = \int f(x) P(x) dx \end{aligned}$$

When $\alpha = \beta = 1/2$

$$P(\theta | \hat{x} = 1) = P(\theta) \frac{\frac{1}{2} \cdot (1-\theta) + \beta \theta}{\int_0^1 (\frac{1}{2} (1-\theta) + \frac{1}{2} \theta) P(\theta) d\theta}$$

$$= P(\theta) \frac{\frac{1}{2}}{\frac{1}{2} \int_0^1 P(\theta) d\theta}$$

$$= P(\theta) \quad \dots \dots \int_0^1 P(\theta) d\theta = 1$$

When $\alpha = \beta = 0$:

$$P(\theta | \hat{x} = 1) = P(\theta) \frac{1-\theta}{\int_0^1 (1-\theta) P(\theta) d\theta}$$

$$= P(\theta) \frac{1-\theta}{E[1-\theta]} \quad \dots \dots E[f(x)] = \int f(x) P(x) dx.$$

(d)

We have $P(\theta | \hat{x} = 1) = P(\theta) \frac{(1-\alpha)(1-\theta) + \beta \theta}{\int_0^1 ((1-\alpha)(1-\theta) + \beta \theta) P(\theta) d\theta}$

... From Q2.c

when $P(\theta) = 1$:

$$P(\theta | \hat{x} = 1) = P(\theta) \cdot \frac{(1-\alpha)(1-\theta) + \beta\theta}{\int_0^1 ((1-\alpha)(1-\theta) + \beta\theta) P(\theta) d\theta}$$

$$= \frac{(1-\alpha)(1-\theta) + \beta\theta}{\int_0^1 ((1-\alpha)(1-\theta) + \beta\theta) d\theta}$$

For $\int_0^1 ((1-\alpha)(1-\theta) + \beta\theta)$:

$$\begin{aligned} \int_0^1 ((1-\alpha)(1-\theta) + \beta\theta) &= \int_0^1 ((1-\theta-\alpha + \alpha\theta) + \beta\theta) d\theta \\ &= \left(\theta - \frac{1}{2}\theta^2 - \alpha\theta + \alpha \frac{\theta^2}{2} + \beta \frac{\theta^2}{2} \right) \Big|_0^1 \\ &= -\frac{1}{2}\alpha + \frac{1}{2}\beta + \frac{1}{2} \end{aligned}$$

Base on $\alpha = \beta$

$$P(\theta | \hat{x} = 1) = \frac{2((1-\alpha)(1-\theta) + \beta\theta)}{-\alpha + \beta + 1}$$

$$= \frac{2((1-\theta-\alpha + \alpha\theta) + \beta\theta)}{1}$$

$$= 2(1-\theta-\alpha + \alpha\theta)$$

$$= 2 - 2\theta - 2\alpha + 4\alpha\theta$$

(e)

$$\begin{array}{l} \alpha = 1 \\ \alpha = \frac{3}{4} \\ \alpha = \frac{1}{2} \\ \alpha = \frac{1}{4} \\ \alpha = 0 \end{array}$$

$$\alpha = 0: P(\theta | \hat{x}=1) = 2 - 2\theta$$

$$\alpha = \frac{1}{4}: P(\theta | \hat{x}=1) = 2 - 2\theta - \frac{1}{2} + \theta = \frac{3}{2} - \theta$$

$$\alpha = \frac{1}{2}: P(\theta | \hat{x}=1) = 2 - 2\theta - 1 + 2\theta = 1$$

$$\alpha = \frac{3}{4}: P(\theta | \hat{x}=1) = 2 - 2\theta - \frac{3}{2} + 3\theta = \frac{1}{2} + \theta$$

$$\alpha = 1: P(\theta | \hat{x}=1) = 2 - 2\theta - 2 + 4\theta = 2\theta$$

Therefore:

1. When $\alpha < 0.5$, the agent result is negative correlation of the coin result.

2. When $\alpha = 0.5$, we can not judge.

3. When $0.5 < \alpha < 1$, the agent result is positive correlation of the coin result.

4. When $\alpha = 1$, the agent result same as coin.

5. When $\alpha = 0$, the agent gets the opposite of coin.

Question 3

(a)

cdf of X :

$$F_X(x) = P(X \leq x)$$

$$= \int_{-\infty}^x P_X(u) du$$

$$= \int_0^x P_X(u) du \quad \dots \dots x \in \mathbb{C}(1]$$

Because X, Y are continuous random variables,

and $X > 0, Y > 0$.

cdf of Y :

$$F_Y(y) = P(Y \leq y)$$

$$= P\left(\frac{1}{X} \leq y\right).$$

$$= P\left(\frac{1}{y} \leq X\right)$$

$$= 1 - P(X \leq \frac{1}{y}) \quad \dots \dots y \in \mathbb{C}(1, +\infty)$$

$$P(X < \frac{1}{y}) = \int_0^1 P_X(n) dn$$

$$\text{Therefore, } F_Y(y) = 1 - \int_0^1 P_X(n) dn$$

$$P_Y(y)$$

$$= F'_Y(y)$$

$$= \frac{d}{dy} \left(1 - \int_0^{\frac{1}{y}} P_X(n) dn \right)$$

$$= - \frac{d}{dy} \left(\int_0^{\frac{1}{y}} P_X(n) dn \right)$$

$$= - \left(\frac{1}{y} \right)' P_X\left(\frac{1}{y}\right)$$

$$= \frac{1}{y^2} P_X\left(\frac{1}{y}\right)$$

$$\dots \frac{d}{dx} \int_a^{u(x)} f(n) dn = u'(x) f(u(x)).$$

(b)

when player wins:

The player wins: m-1

The probability: $\int_m^{\infty} P_Y(y) dy$

when player lose:

The player loss: 1

The probability: $\int_1^m P_Y(y) dy$.

Then,

$$\begin{aligned} E &= (m-1) \int_m^{\infty} P_Y(y) dy - \int_1^m P_Y(y) dy \\ &= m \int_m^{\infty} P_Y(y) dy - \int_1^m P_Y(y) dy - \int_m^{\infty} P_Y(y) dy \\ &= m \int_m^{\infty} P_Y(y) dy - \int_1^{\infty} P_Y(y) dy \end{aligned}$$